

Testing Sample-Based Accounts of Probability Judgements Using a Ranking Task

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STATISTICAL MODELING in PSYCHOLOGY
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Introduction

- People's explicit probability judgements often appear to be probabilistically incoherent (i.e., Linda problem)
- Biases in probability judgements can arise from rational reasoning processes based on mental samples
- Sample-based probability judgement models assume that...
 - ... samples generated from probability distributions in mind
 - ... each sample takes time to generate
 - ... infinite number of samples result in unbiased judgements

One sample-based model: PT+N model

Core assumptions in Probability Theory Plus Noise (PT+N) model (Costello & Watts, 2014):

- DM holds coherent sets of beliefs such that $P(a) + P(\neg a) = 1$
- When asked to explicitly estimate $P(A)$...
 - ... DM requests N samples, according to **a fixed belief**, $P(a)$
 - ... DM counts number of critical samples i that indicate A .
 - ... sample has some chance d of being mistaken for its complement

\Rightarrow estimation based on relative frequencies of number of critical samples

One sample-based model: PT+N model

From these assumptions,

- Expected value of individual estimates (Costello & Watts, 2014, Eq. 3)

$$P_E(A) = (1 - 2d)P(a) + d$$

- Estimate of number of critical samples (Howe & Costello, 2020, Eq. 4)

$$P(i|(N, p) = \text{Bin}(N, p) = \binom{N}{i} p^i (1 - p)^{N-i}$$

$$p = P_E(A)$$

N : total sample size; i : the number of critical events

The limitation

- PT+N account only tested in estimation tasks
 - Mono-operation bias
 - Sampling parameters not identifiable without additional assumptions
- A novel ranking task is proposed
 - Ranking task
 - PT+N rank model
 - Derive qualitative predictions
 - Test qualitative predictions experimentally

The ranking Task

Please rank the four events below according to their likelihoods.

Assign **rank 1 to the event that you think is most likely to occur**, rank 2 to the event that you think is second most likely to occur, and so forth. Thus the event that you think is least likely to occur should receive rank 4. **Please specify the rank by selecting the corresponding number on the right.**

	1	2	3	4
On a random day in England, it will not be rainy.	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will be cold.	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will be rainy.	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will not be cold.	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>

Normative rules in ranking task

- The complement rule requires $P(a) + P(\neg a) = 1$
 - For logically possible rankings,
 $\text{Rank}(A) + \text{Rank}(\neg A) = \text{Rank}(B) + \text{Rank}(\neg B)$
- 75 possible ranks over four events
 - **17 of them are logically possible**
 - Rankings with vs. without ties
 - possibility to categorise logically impossible rankings

Normative rules in ranking task

Logically possible rankings. . .

- With ties

i.e., $\text{Rank}(A) + \text{Rank}(\neg A) = 4$; $\text{Rank}(B) + \text{Rank}(\neg B) = 4$

	1	2	3	4
On a random day in England, it will be rainy.	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will not be rainy.	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>
On a random day in England, it will not be cold.	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will be cold.	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>

Normative rules in ranking task

Logically possible rankings. . .

- Without ties

i.e., $\text{Rank}(A) + \text{Rank}(\neg A) = 5$; $\text{Rank}(B) + \text{Rank}(\neg B) = 5$

	1	2	3	4
On a random day in England, it will be rainy.	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will not be rainy.	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>
On a random day in England, it will not be cold.	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>
On a random day in England, it will be cold.	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>

Normative rules in ranking task

Categorise rankings without ties. . .

- class one illogical rankings:
i.e., $A \ \& \ \neg A$ simultaneously ranked above $B \ \& \ \neg B$

	1	2	3	4
On a random day in England, it will be rainy.	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will not be rainy.	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will not be cold.	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>
On a random day in England, it will be cold.	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>

Normative rules in ranking task

Categorise rankings without ties...

- class two illogical rankings:
i.e., $A > B$ & $\neg A > \neg B$ simultaneously hold in one response

	1	2	3	4
On a random day in England, it will be rainy.	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will not be rainy.	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>
On a random day in England, it will not be cold.	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>
On a random day in England, it will be cold.	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>

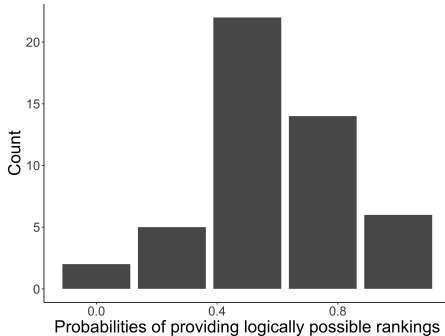
Normative rules in ranking task

Categorise rankings without ties. . .

- logically possible rankings

	1	2	3	4
On a random day in England, it will be rainy.	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
On a random day in England, it will not be rainy.	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>
On a random day in England, it will not be cold.	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>
On a random day in England, it will be cold.	<input type="radio"/>	<input checked="" type="radio"/>	<input type="radio"/>	<input type="radio"/>

A new bias?



49 participants were asked to rank $\{\text{event } A, \text{event } \neg A, \text{event } B, \text{event } \neg B\}$.

On average, around two out of the four rankings that participants provided were logically possible (mean = 0.59, SD = 0.24).

A new bias?

Is it a real bias, or merely the consequence of mental sampling processes?

A not so clear answer: PT+N can produce logically impossible rankings. But we need to test its sufficiency in explaining rankings.

PT+N rank model

What does PT+N model imply the rank-based probability judgements?

PT+N rank model:

- model assumptions:
 - DM holds coherent sets of beliefs
 - When asked to rank a set of events (i.e., $\{A, \neg A, B, \neg B\}$)...
 - ... DM implements direct sampling for each event in set
 - ... **for events in same set, one uses same sample size**
 - ... **derives ranks by comparing number of critical samples**
 - sample has some chance d of being mistaken for its complement

PT+N rank model

Analytic formula for rankings without any ties:

$$P(A > B > C > D) = P(C_A > C_B > C_C > C_D)$$

$$= \sum_{i_d=0}^{N-3} f(i_d, N, d) \sum_{i_c=(i_d+1)}^{N-2} f(i_c, N, c) \sum_{i_b=(i_c+1)}^{N-1} f(i_b, N, b) \sum_{i_a=(i_b+1)}^N f(i_a, N, a)$$

$$f(i, N, p) = \binom{i}{N} p^i (1-p)^{N-i}$$

N: total sample size; i: the number of critical event; p: the expected value of P(A)

Assume $N = 4$:

i_A	i_B	i_C	i_D
3	2	1	0
4	2	1	0
4	3	1	0
4	3	2	0
4	3	2	1

- Substitute A, B, C, D with A, $\neg A$, B, $\neg B$
in 24 linear orderings

$\{A, \neg A, B, \neg B\}$

$\{A, \neg A, \neg B, B\}$

$\{A, B, \neg A, \neg B\}$

.....

Predictions for rankings

Analytic formula for ranking with ties:

$$\begin{aligned}P(A > B > C = D) \\&= \sum_{i_{c,d}=0}^{N-2} f(i_{c,d}, N, d) f(i_{c,d}, N, c) \sum_{i_b=(i_{c,d}+1)}^{N-1} f(i_b, N, b) \sum_{i_a=(i_b+1)}^N f(i_a, N, a) \\P(A > B = C > D) &= \dots \\P(A = B > C > D) &= \dots\end{aligned}$$

$$\begin{aligned}P(A = B > C = D) \\&= \sum_{i_{c,d}=0}^{N-1} f(i_{c,d}, d) f(i_{c,d}, n, c) \sum_{i_{a,b}=(i_{c,d}+1)}^N f(i_{a,b}, b) f(i_{a,b}, a)\end{aligned}$$

$$\begin{aligned}P(A = B = C > D) \\&= \sum_{i_d=0}^{N-1} f(i_d, N, d) \sum_{i_{a,b,c}=(i_d+1)}^N f(i_{a,b,c}, N, c) f(i_{a,b,c}, N, b) f(i_{a,b,c}, N, a) \\P(A > B = C = D) &= \dots\end{aligned}$$

$$\begin{aligned}P(A = B = C = D) \\&= \sum_{i_{a,b,c,d}=0}^N f(i_{a,b,c,d}, N, d) f(i_{a,b,c,d}, N, c) f(i_{a,b,c,d}, N, b) f(i_{a,b,c,d}, N, a)\end{aligned}$$

An empirical issue

Will tied rankings cause troubles on model testing?

We hypothesise that if ties are allowed, lazy participants will overuse ties.

The solution

- add additional assumptions:
 - ...
 - ...
 - ...
 - **When tied rankings not allowed, participants will produce a random linear ordering among tied events but still follow the ordering among the untied events**

The solution

On a random day in England, it will be rainy.



On a random day in England, it will **not** be rainy.



On a random day in England, it will **not** be cold.



On a random day in England, it will be cold.



Derive qualitative predictions

- How do we derive the qualitative predictions?
 - Four parameters enter the model, N , d , $A(\text{or } \neg A)$, B (or $\neg B$).
 - There categories of the responses
 - class one illogical rankings
 - class two illogical rankings
 - logically possible rankings

Derive qualitative predictions

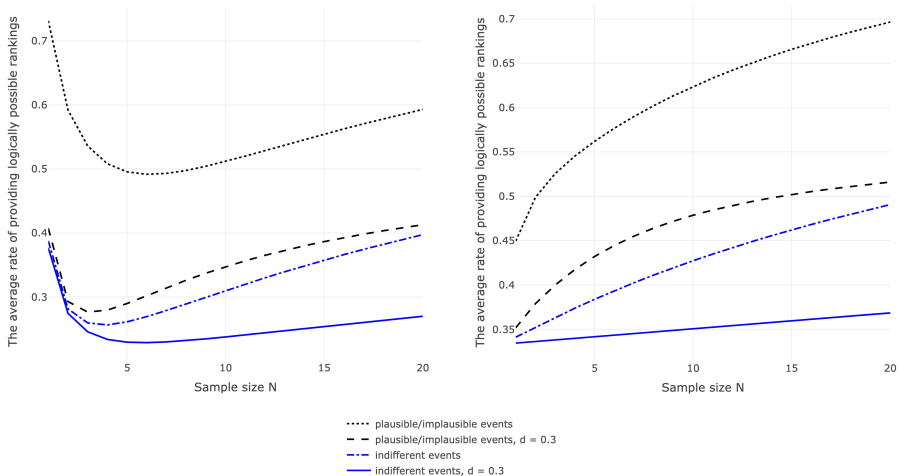
Simulation

- Inputs: different parameter values
 - Inherent beliefs (i.e., $P(a)$)
 - beliefs about plausible/improbable events, drawn from Beta(1, 10)
 - beliefs about indifferent events, drawn from Beta(10, 10)
 - reading error d
 - $d = 0$ or 0.3
 - sample size N
 - range from 1 to 20 in a step of 1
- Repeat 100000 times
- Results
 - simulate ranking data for two types of event sets
 - plausible/improbable vs. indifferent event sets
 - calculate prob. of providing three categories of rankings

Derive qualitative predictions

Average rate of producing logically possible rankings

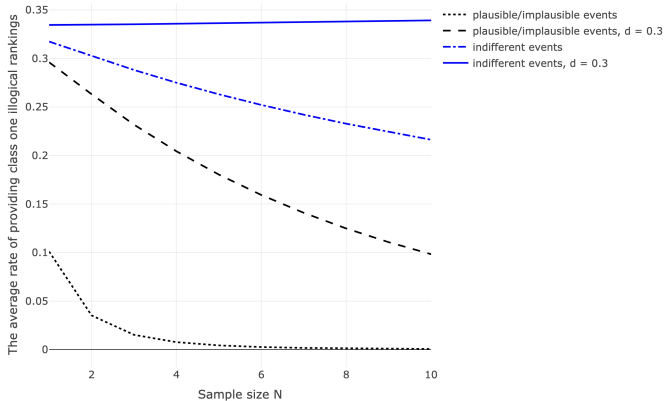
left: allow ties vs. right: not allow ties



Derive qualitative predictions

Average rate of producing class one illogical rankings

ties not allowed



Testing the qualitative predictions experimentally

Design

- manipulate inherent beliefs within subjects
 - ask participants to rank two types of events
plausible/improbable vs. indifferent event set
- manipulate whether or not ties are allowed between subjects

Hypothesis

If we compare plausible/improbable event sets to indifferent event sets,

- Average rate of providing class one illogical rankings will be lower
- Average rate of providing logically possible rankings will be higher

Summary

- Ranking task to investigate coherence of beliefs in a novel way
 - Biases in rank-based probability judgements
 - can be explained by existing sample-based models
- Need to understand and test the consequences of sampling
 - Ranking tasks allow us to do so
 - testing qualitative predictions of rankings

Discussion

- Future research ideas
 - Model is identifiable
 - ⇒ parameters can be estimated on individual level
 - Manipulate sample size N by...
 - ... manipulating response time
 - ... manipulating the structure of sentences
 - A randomly selected resident is living in southern Germany.
 - There are more people living in southern Germany than in northern Germany.

Thank you!

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